

Can CO₂ Turbulent Flux Be Measured by Lidar? A Preliminary Study

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(Manuscript received 19 January 2010, in final form 15 April 2010)

ABSTRACT

The vertical profiling of CO₂ turbulent fluxes in the atmospheric boundary layer (ABL) is investigated using a coherent differential absorption lidar (CDIAL) operated nearby a tall tower in Wisconsin during June 2007. A CDIAL can perform simultaneous range-resolved CO₂ DIAL and velocity measurements. The lidar eddy covariance technique is presented. The aims of the study are (i) an assessment of performance and current limitation of available CDIAL for CO₂ turbulent fluxes and (ii) the derivation of instrument specifications to build a future CDIAL to perform accurate range-resolved CO₂ fluxes. Experimental lidar CO₂ mixing ratio and vertical velocity profiles are successfully compared with in situ sensors measurements. Time and space integral scales of turbulence in the ABL are addressed that result in limitation for time averaging and range accumulation. A first attempt to infer CO₂ fluxes using an eddy covariance technique with currently available 2- μ m CDIAL dataset is reported.

1. Introduction

The study of transport processes by turbulent fluxes is a key to understanding the exchanges that take place among the various components of the biosphere (i.e., the surface including vegetation and the atmosphere) at different time and space scales. For the past several decades, turbulent flux measurements of scalars have been conducted with in situ probes by using the eddy covariance (EC) technique (Desai et al. 2005). This results in a rather high-accuracy measurement ($\approx 10\%$) but with an inherent representativeness limitation to small or local scales. Furthermore, considering the need for large scales both in vertical and horizontal dimensions, it calls for networks of tall towers. Even if the usual height is limited to tens of meters, their deployment is scarce and expensive. The few tall towers that extend several hundreds of meters are suited for vertical sampling in the nocturnal layer (NL) but fail short to sample the daily convective atmospheric boundary layer (CBL) or nocturnal residual layer (RL) with

height $h \geq 0.5$ km. Only the deployment of instrumented aircraft enables one to profile the turbulent fluxes but with the limitation of requiring dedicated field campaigns.

Having these restrictions in mind, and even knowing that footprint models may enlarge the in situ probe measurement representativeness (Wang et al. 2006), we think it is worth addressing the capability of remote sensors to profile turbulent fluxes in either vertical or horizontal dimensions. To this end, lidar techniques [i.e., elastic backscatter, differential absorption lidar (DIAL), Raman, Doppler] have shown a great capability to profile aerosol particles, minor gas concentration, and wind velocity along the lidar line of sight or in 2D or 3D. Through combining two of these capabilities (gas concentration and wind velocity), lidars have the potential to make range-resolved flux measurements using an eddy covariance method. Preliminary studies have been conducted using a combination of ground-based (Giez et al. 1999) or airborne (Kiemle et al. 2007) Doppler and DIAL lidars, or aerosol flux using ground-based Doppler and Raman lidars (Engelmann et al. 2008). Moreover, the flexibility of the lidar technique in time averaging and over range accumulation is well suited to address integral scale issues up to several kilometers.

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Presently, we are interested in an application of lidar for CO₂ fluxes using the eddy covariance technique. Instead of using two different lidars, one for concentration and one for velocity measurements (see Giez et al. 1999; Kiemle et al. 2007), we propose to use one single 2- μ m coherent differential absorption lidar (CDIAL) for simultaneous measurements of CO₂ concentration and velocity (Gibert et al. 2006). However, one limitation is it requires a large number of independent samples for accurate CO₂ concentration measurements. The signal-to-noise ratio (SNR) in coherent (or heterodyne) detection on a single-shot basis is limited to about unity because of the speckle noise. Single-shot SNR can be improved using a speckle diversity technique (Favreau et al. 2000) but still requires a large number of samples to achieve high SNR [$\approx(10^2\text{--}10^3)$] as required for 1% accuracy on absorption optical depth to derive high-accuracy molecular density and mixing ratio. The 2- μ m CDIAL systems demonstrated a precision of $\sim 1\%$ for CO₂ mixing ratio for an accumulation time of ~ 30 min and range resolution of ~ 1 km (Koch et al. 2008; Gibert et al. 2008).

The goal of the present paper is (i) to assess the performance and current limitation for CO₂ flux measurements using one single DIAL and Doppler lidar and (ii) to derive the specifications of a future instrument to retrieve accurate CO₂ fluxes. Section 2 presents the experimental site in Wisconsin with the different instruments (i.e., an instrumented 447-m tower and a ground-based 2- μ m CDIAL) and the CO₂ flux measurement technique. Section 3 presents the meteorological conditions during the 3-day period (i.e., 14–16 June 2007) devoted to field measurements. The 2- μ m CDIAL for simultaneous CO₂ and velocity measurements is presented in section 4. Section 5 deals with CO₂ and vertical velocity fluctuations as inferred by in situ sensors. Section 6 addresses issues of time and space integral scales as reported by Lenschow and Stankov (1986) that result in limitation for time averaging and range accumulation. This process allows deriving the upper bounds to avoid biases in CO₂ flux estimates using a 2- μ m CDIAL. Section 7 is a first attempt to infer CO₂ fluxes using an eddy covariance technique and current 2- μ m CDIAL dataset. Finally, section 8 deals with the specifications for a future CDIAL to perform accurate ($<50\%$) CO₂ flux measurements.

2. Experimental site, instrumentation, and lidar methodology

a. Instrumentation

The National Aeronautics and Space Administration (NASA) Langley coherent 2- μ m DIAL was positioned underneath the 447-m WLEF tall tower (45.945°N, 90.272°W) in Park Falls, Wisconsin, approximately 40 m

away from the tower's centerline. Two in situ infrared gas analyzers (IRGAs; LiCor Model Li-6251) provide 2-min mean CO₂ mixing ratios at six levels above the ground (11, 30, 76, 122, 244, and 396 m). Turbulent winds, virtual potential temperature, and H₂O mixing ratio are also measured by three sonic anemometers and other IRGAs at three levels: 30, 122, and 396 m. A ground-based meteorological station also provides net radiation and surface pressure, temperature, and moisture (Berger et al. 2001).

The 2- μ m CDIAL transmitter is a 90-mJ, 140-ns, 5-Hz pulsed Ho, Tm:LuLiF seeded oscillator described in Koch et al. (2004, 2007). The online wavelength of the transmitter is locked onto the side of the R22 CO₂ absorption line at 2053.204 nm with a frequency stability of 1.9 MHz, which is needed for unbiased measurements. However, the offline was positioned 0.25 nm away on weak H₂O absorption lines, which results in bias that can be corrected as discussed in section 4. The lidar beam can be scanned to probe the atmosphere in 3D or vertically pointing at zenith or horizontally. The heterodyne RF signals are digitized on 8 bits at a 500-MHz sampling frequency and stored on a PC. Postprocessing is conducted using both power and velocity estimators (Gibert et al. 2006). The atmospheric return signals are processed range gate by range gate and then accumulated for N shots. The time and range resolutions are 40 s and 75 m, respectively. As presented in section 1, the vertical velocity and CO₂ profiles are of first interest. In addition, horizontal wind measurements were performed every 30 min.

b. Simultaneous vertical structure, vertical velocity, CO₂ mixing ratio, and turbulent flux estimates by 2- μ m CDIAL

The height of the convective and the residual (RL) boundary layer are calculated using the second derivative of the backscatter signal profile (Menut et al. 1999). Given the lidar limitation at short range (150 m), the lidar was able to probe the residual layer at night and most of the convective boundary layer during daytime. The nocturnal boundary layer height (NL) was estimated using the second derivative of CO₂ mixing ratio profiles as provided by the WLEF radio tower.

The vertical (or horizontal) velocity component on the lidar line of sight is computed from the mean Doppler frequency shift $\Delta\nu$ due to particle motion in Δz :

$$w = -\lambda \Delta\nu / 2. \quad (1)$$

An estimate of a mean CO₂ mixing ratio at $z + \Delta z/2$ is obtained assuming that the on- and off-line probing frequencies are close enough that backscatter and extinction coefficient differences are negligible:

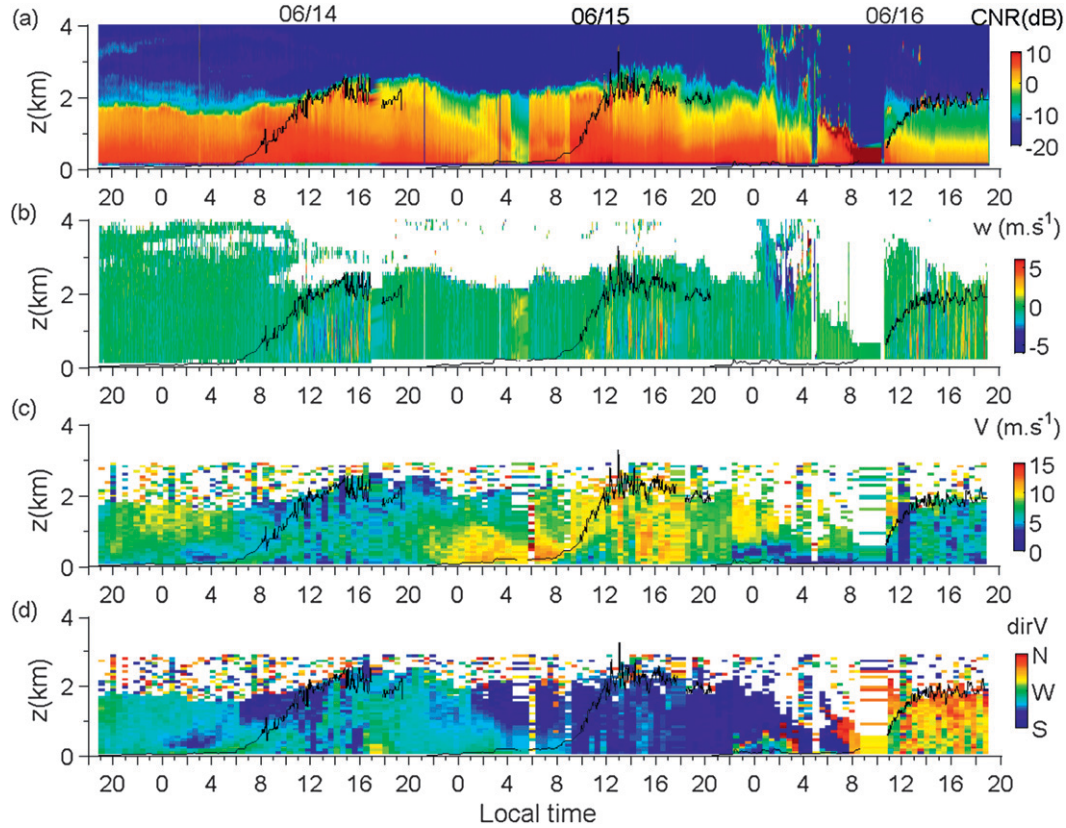


FIG. 1. (a) The off-line carrier-to-noise ratio (CNR), (b) vertical velocity (w), (c) horizontal wind speed (V), and (d) direction (dirV) as a function of the local time. The ABL height is indicated with a black solid line.

$$\rho_{\text{CO}_2} = \frac{1}{2n_{\text{air}}\Delta\tilde{\sigma}_{\text{CO}_2}} \ln \left[\frac{P_{\text{off}}(z + \Delta z)P_{\text{on}}(z)}{P_{\text{off}}(z)P_{\text{on}}(z + \Delta z)} \right] + C + D, \quad (2)$$

where $C = -\rho_{\text{H}_2\text{O}}\Delta\tilde{\sigma}_{\text{H}_2\text{O}}/\Delta\tilde{\sigma}_{\text{CO}_2}$ is a corrective term due to the overlap of water vapor absorption line on the off laser line, $\rho_{\text{H}_2\text{O}}$ is the mean water vapor mixing ratio, $\Delta\tilde{\sigma}$ is the differential cross section, n_{air} is the dry air density, and $D = [\text{SNR}_{\text{on}}^{-2}(z) - \text{SNR}_{\text{off}}^{-2}(z) - \text{SNR}_{\text{on}}^{-2}(z + \Delta z) + \text{SNR}_{\text{off}}^{-2}(z + \Delta z)]/(4n_{\text{air}}\Delta\tilde{\sigma}_{\text{CO}_2})$ is a corrective term due to DIAL equation nonlinearity and depends on the off and on SNR [from appendix C in Gibert et al. (2008)].

The differential cross section and dry-air density are computed using the most accurate spectroscopic data (Toth et al. 2006, 2007) and on-site data from the WLEF tower in situ sensors for temperature, pressure, and specific humidity. However, the meteorological sensors measurements are extrapolated up to 3 km assuming a linear decrease of temperature and an exponential decrease of pressure. Note that $\rho_{\text{H}_2\text{O}}$ is assumed to be constant in the atmospheric boundary layer (ABL) above $z > 396$ m and negligible in the free troposphere.

To infer a CO_2 flux estimate using the eddy covariance method, we need high-frequency measurements of CO_2 and velocities. The key question about sampling frequency is presented in section 6. To avoid any bias in CO_2 flux, the lidar time and range resolutions need to be smaller than the vertical and horizontal scales of turbulence. The CO_2 EC flux relies on correlation between the fluctuations of CO_2 mixing ratio (ρ'_{CO_2}) and vertical velocities (w') due to turbulence only:

$$F_{\text{CO}_2} = \overline{\langle w' \rangle' \langle \rho'_{\text{CO}_2} \rangle'} + F_C, \quad (3)$$

where the angle brackets and overbar stand for range accumulation and time averaging. As before, $F_C = -F_{\text{H}_2\text{O}}\Delta\tilde{\sigma}_{\text{H}_2\text{O}}/\Delta\tilde{\sigma}_{\text{CO}_2}$ is a corrective flux term due to partial water vapor absorption in the off-line mode.

3. Case study

The field experiments were conducted on three days: 14–16 June 2007, with different synoptic conditions (Fig. 1).

The sun rises at 0530 and sets at 2030 local time; 14 and 16 June (D14 and D16 hereafter) are characterized by weak wind speed conditions with $V < 5 \text{ m s}^{-1}$ whereas for 15 June (D15) $V = 10 \text{ m s}^{-1}$. Cumulus clouds were present during daytime and large entrainment zones (especially for D14 and D15) were observed at the top of the convective boundary layer. It results in large variations of lidar carrier-to-noise ratio (CNR). The 14 June night (N14, between 13 and 14 June) is relatively steady with weak wind conditions close to the surface that increase with height following a lognormal law profile. The 15 June night (N15) is characterized by a strong nocturnal jet with $V > 10 \text{ m s}^{-1}$ for $z > 0.1 \text{ km}$ up to 1 km. It is worth noticing that despite low lidar CNR around 0500 local time, we can identify downward slanting aerosols structures that led us to suspect significant subsidence motion during this night. During the night of 16 June (N16), a thunderstorm resulted in some rain around 0300 LT. Strong updrafts and downdrafts were identified in front of it (at 2200 and 0100 LT, respectively) and during the rain period (at 0200 and 0300 LT), which provided additional sources of turbulence during this night.

4. CDIAL performances

a. Vertical velocity

Following Frehlich et al. (1998), the power spectra of lidar or in situ vertical velocity (w) is

$$\Phi_w(\nu) = 2 \frac{\Delta t}{M} |\text{DFT}_w(\nu/\Delta\nu)|^2, \quad (4)$$

where DFT is the discrete Fourier transform, Δt is the time interval between the estimates, [i.e., 40 s (lidar) and 0.2 s (in situ)], $\Delta\nu = 1/(N\Delta t)$ is the frequency resolution, N is the number of independent measurements, and M is the number of vertical velocity estimates in a given time gate.

Figure 2 shows a comparison of velocity spectrum for lidar and in situ sensors for nighttime and daytime. Figures 2a–c show that the spectra are in good agreement

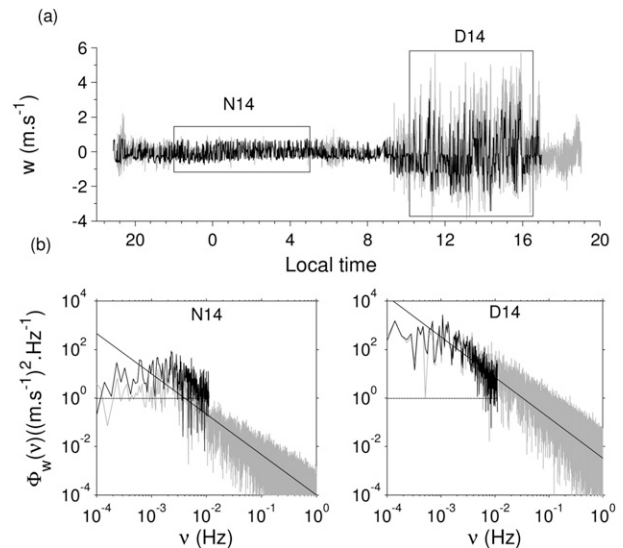


FIG. 2. Vertical velocity time series for (a) the sonic anemometer at 396 m and (b) the CDIAL at 375 m on 14 June. The corresponding spectra for the night N14 and day D14 are displayed. The gray line indicates the in situ sensor and the black line is for the lidar. The usual Kolmogorov $-5/3$ slope is indicated. The black dashed line represents the noise level.

in the CBL. The variance from lidar measurements seems slightly reduced, though (Fig. 2a). In the residual layer (nighttime), the natural variance of vertical velocity is in the same order of magnitude as the lidar noise variance. However, a turbulent spectrum peaking at $3 \times 10^{-3} \text{ Hz}$ with a $-5/3$ slope in the inertial subrange is still apparent. From the spectra, we estimate instrumental noise and natural vertical velocity standard deviations (Frehlich et al. 1998; Gibert et al. 2007a), as reported in Table 1.

b. CO_2 mixing ratio

To make a relevant comparison between 2- μm CDIAL and in situ CO_2 mixing ratio we average lidar estimates during 6 h for N14 and D14. As displayed in Fig. 3, it enables us to make for the first time a comparison of CO_2 profiles between a 2- μm CDIAL and in situ

TABLE 1. Standard deviation of vertical velocity and CO_2 mixing ratio measurements. In situ data are from the 396-m high level of the WLEF tower. Lidar data are at 375 m. The CO_2 lidar measurement standard deviation is supposed to equal the noise variance. Time periods in 2007 are 14 Jun [N14 (2200–0500), D14 (1030–1630)]; 15 Jun [N15 (2200–0400), D15 (1100–1700)]; and 16 Jun [N16 (2100–0300), D16 (1130–1730)].

Time period		N14	D14	N15	D15	N16	D16
$\sigma(w)$ (m s^{-1})	In situ	0.15	1.14	0.05	1.18	0.27	1.02
	Lidar	0.36 ± 0.13	1.05 ± 0.13	0.13 ± 0.07	0.91 ± 0.07	0.40 ± 0.14	1.04 ± 0.14
$\sigma(\rho_{\text{CO}_2})$ (ppm)	In situ	0.25 ± 0.06	0.43 ± 0.06	0.31 ± 0.06	0.45 ± 0.06	0.37 ± 0.06	0.42 ± 0.06
	Lidar						
	75 m	± 842	± 792	± 884	± 810	± 820	± 970
	1 km	± 34	± 38	± 45	± 45	± 38	± 48

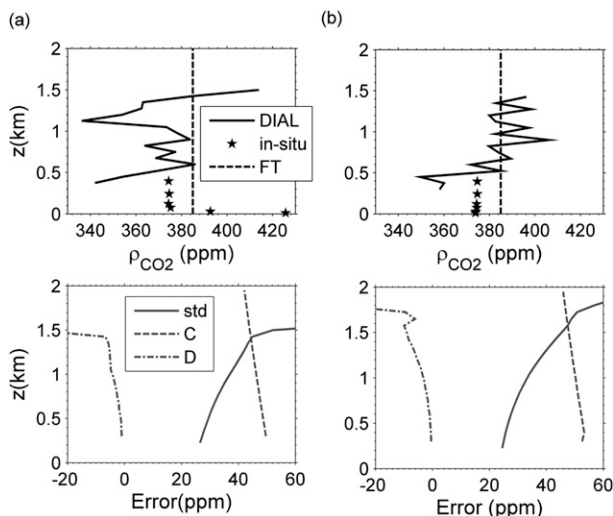


FIG. 3. The 6-h, 150-m averaged CDIAL CO₂ mixing ratio profiles for (a) N14 and (b) D14 compared with the in situ profile up to 396 m. The free troposphere (FT) mixing ratio is indicated with a dashed line. Corresponding (c) CDIAL standard deviation (std dev) and (d) bias profiles.

sensors up to 396 m, the maximum height of the tower. Biases and standard deviation profiles have been calculated as well. The bias term, C , due to water vapor absorption [see Eq. (2)] is rather constant along the vertical according to an assumption of constant mixing ratio in the residual or convective boundary layers. The bias term D and standard deviation increase with altitude as a result of a decrease of the Online SNR due to CO₂ absorption.

After correction for potential biases, the N14 lidar CO₂ profile agrees well with in situ data. For D14, the lidar CO₂ mixing ratios increase with height. Notice that airborne CO₂ measurements on 11 June 2007, conducted by the National Oceanic and Atmospheric Administration (NOAA) around the WLEF site ($46.00^\circ \pm 0.05^\circ\text{N}$, $90.17^\circ \pm 0.03^\circ\text{W}$) reported a free troposphere CO₂ mixing ratio of (384.5 ± 0.4) ppm from 2200 m up to 3900 m. Such comparable values for CO₂ mixing ratio in the CBL by lidar and free troposphere by aircraft can be explained by a thick and an active entrainment zone on D14 as displayed in Fig. 1a.

At time and range resolutions of 40 s and 75 m, respectively, a white noise prevails in lidar measurements of CO₂ mixing ratio (Table 1). The error in CO₂ measurements is slightly larger than what was estimated previously (Gibert et al. 2008; see the standard deviation for resolutions of 1 km and 2 min). It is mainly due to a lower CO₂ absorption (i.e., the on-line was moved to the edge of the CO₂ absorption line in order to increase the range of measurements). The standard deviation decreases as the square root of time and range averaging as expected. The online laser frequency jitter of 1.9 MHz contributes to 0.2 ppm to standard deviation for a 40-s averaging. Other

uncertainties and biases as quantified in Gibert et al. (2008) are negligible.

5. Turbulent fluctuations of CO₂ mixing ratio and vertical velocity in the ABL from in situ sensors

To derive the incidence of turbulent fluctuations in CO₂ flux retrievals by lidar, a necessary step is an estimate of the relevant space and time integral scales (see section 6). Here the turbulent fluctuations are analyzed using high-time-resolution and high-accuracy in situ probes. The standard deviation of vertical velocity is driven by surface heat flux variations. It is the largest in the middle of the CBL following the similarity law as displayed in Fig. 4. Mean velocity fluctuations are around $\pm 1 \text{ m s}^{-1}$. During the three nights, the vertical velocity fluctuations at 396 m can be as low as $\pm 0.03 \text{ m s}^{-1}$ (N15 case), with large variations from one hour to another (not shown here), and can reach $\pm 0.5 \text{ m s}^{-1}$ (N16 case). The daytime CO₂ mixing ratio fluctuations are rather similar from day to day, with a value around ± 0.5 ppm. Large CO₂ variations are seen during the morning transition to be up to 8 ppm at 122 m and 3 ppm at 396 m (D16). These are due to the rising CBL and former nocturnal layer into the residual layer. In situ sensors at 396 m are in the residual layer during nighttime and CO₂ mixing ratio turbulent fluctuations are usually as low as ± 0.2 ppm with frequent increases up to ± 0.5 ppm associated with either a nocturnal jet on N15 or a storm on N16. The similarity law for scalars predicts rather similar turbulent fluctuations in the whole convective and residual layers. In comparison to water vapor turbulent fluctuations ($\sim 5\%$ of the mean value; Giez et al. 1999), it is clear that CO₂ fluctuations ($\sim 0.1\%$) are very hard to reach for lidar today, even considering averaging over few hours and/or accumulation over longer ranges (1 km or so). However, the instrumental challenge for coherent lidar is in vertical velocity measurements at night when fluctuations as low as 0.03 m s^{-1} are observed. To achieve such a low figure, time averaging and range accumulation are required for lidar but some limitations are set by turbulence integral scale as discussed in section 6.

6. Space and time integral scales

Time and space scales of turbulence are calculated using time and space covariance of vertical velocity and CO₂ mixing ratio following Lenschow and Stankov (1986) (appendix B).

a. Horizontal or time integral scale

Figure 5a shows hourly time integral scale of vertical velocities $l_w^{(t)}$ from sonic anemometer and 2- μm CDIAL data. During daytime for in situ sensor at 396 m, $l_w^{(t)}$

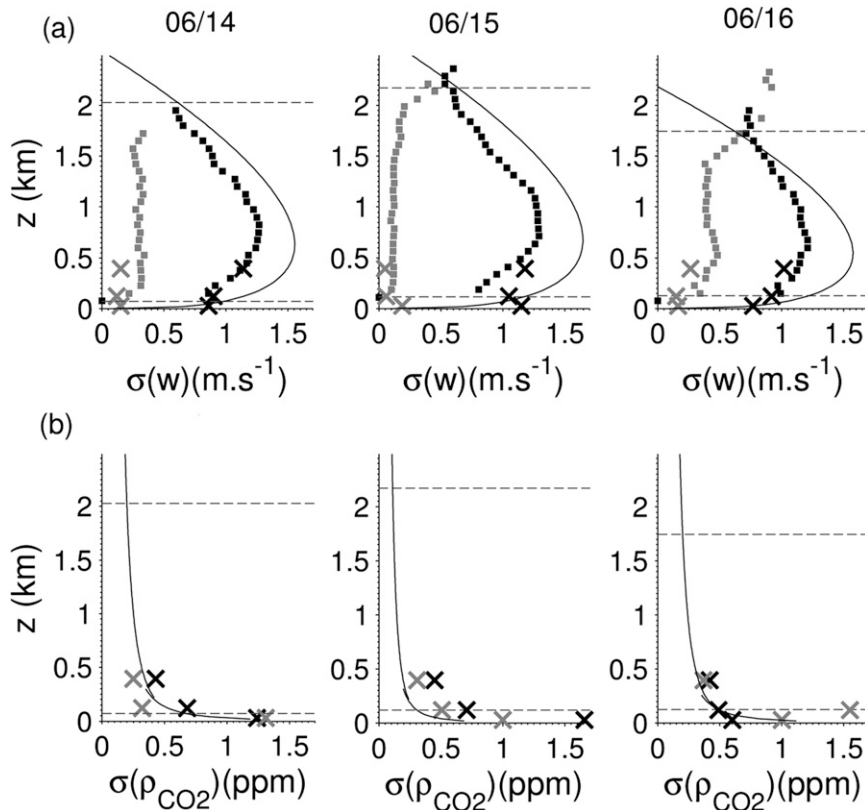


FIG. 4. (a) Vertical velocity and (b) CO₂ mixing ratio natural standard deviation for the CDIAL (squares) and in situ sensors (crosses) for (left to right) 14, 15, and 16 June. Black and gray colors are for day and night, respectively. Theoretical laws from similarity are also displayed (black solid lines). Mean NL and CBL heights (dashed lines) are also displayed.

ranges between 30 and 90 s. Comparison of lidar and in situ estimates are in good agreement. During nighttime large variations of $l_w^{(t)}$ occur. On N15, when a nocturnal jet occurred, $l_w^{(t)}$ ranges between 50 and 150 s, whereas on N14 $l_w^{(t)}$ is lower than 50 s. Close to the surface, at 30 m, a diurnal cycle of $l_w^{(t)}$ can be seen. In the nocturnal boundary layer, the size of turbulent eddies is indeed much reduced by static stability.

Similar integral scales or coherent time of turbulence can be calculated for CO₂ mixing ratio and for CO₂ fluxes (appendix C). Note that $l_{\rho_{CO_2}}^{(t)}$ is larger than $l_w^{(t)}$, especially during the night. The fluctuations of a scalar are linked not only to vertical motion but also to advection. During the day, $l_w^{(t)}$, $l_{\rho_{CO_2}}^{(t)}$, and $l_{F_{CO_2}}^{(t)}$ are comparable, meaning that buoyancy drives both velocity and scalar fluctuations in the CBL as expected.

In Fig. 6a, we analyze the vertical variations of the horizontal integral scale. Above 396 m and during daytime, $l_w^{(t)}$ usually increases with height and reaches a maximum in the middle of the CBL where thermals can be clearly defined (Kaimal et al. 1976; Lenschow et al. 1980; Gibert et al. 2007a). Using vertical profile of horizontal wind from lidar, our results confirm the results that for

$z < z_i/2$, $l_w^{(t)}$ increases with height according to $\sim (zz_i)^{1/2}$. For $z > z_i/2$, $l_w^{(t)}$ decreases significantly with height when the thermals contours are not well delimited and cloud effects and entrainment are important near the CBL top. The $l_w^{(t)}$ profiles at nighttime seem to depend on horizontal wind speed profiles in the residual layer. An increase of the wind speed entails larger structures of turbulence. For example, the nocturnal jet during N15 results in the largest $l_w^{(t)}$ values in our dataset (Figs. 5a and 6a).

b. Vertical integral scale

We use the lidar vertical velocity profiles to estimate the vertical integral scale of turbulence $l_w^{(z)}$. Figure 5b shows that at 375 m in the CBL $l_w^{(z)}$ ranges between 100 and 250 m. These values are smaller than those proposed in Lothon et al. (2006) but are in good agreement with Giez et al. (1999). At night, the integral scale is around 100 m, except for N15. Large $l_w^{(z)}$ values are computed for N16 because of storm downdrafts and updrafts.

Figure 6b shows that during daytime, $l_w^{(z)}$ is maximum in the middle of the CBL, as is $l_w^{(t)}$. Near the CBL top, $l_w^{(z)}$ decreases significantly with height, in agreement with

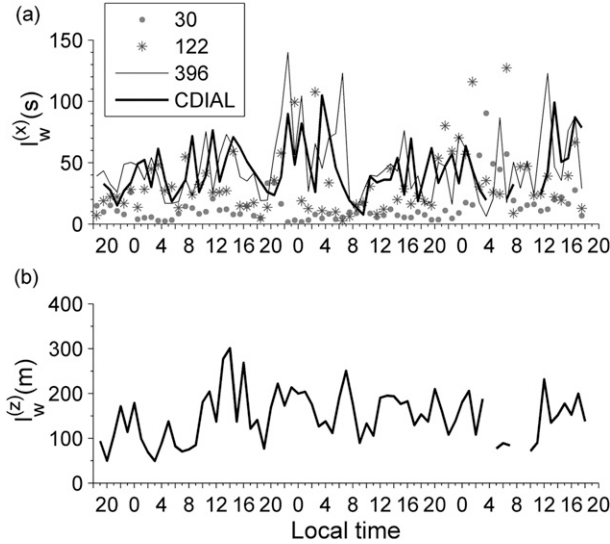


FIG. 5. (a) Hourly time integral scale of vertical velocities $l_w^{(t)}$ using sonic anemometer at 30 (circles), 122 (stars), and 396 m (thin solid line) and CDIAL data at 375 m (thick solid line). (b) Hourly vertical integral scale of vertical velocities $l_w^{(z)}$ at $z = 1$ km. CDIAL vertical velocity resolution is 75 m.

both compressed eddies (Kristensen et al. 1989; Lothon et al. 2006) and decrease of large-scale coherence near the entrainment zone. A decrease of $l_w^{(z)}$ happens also when z decreases. This result is in disagreement with Lothon et al., who found constant or increasing $l_w^{(z)}$ for decreasing z . Our results are consistent with an increase of rather small-scale turbulent structures closer to the ground. During the night and above 396 m, $l_w^{(z)}$ seems to increase with the wind speed (N14) or subsidence large-scale structures (N15 and N16). For the following parts of this paper and for $l_{F_{CO_2}}^{(t)}$, we assume that the flux vertical integral scale is nearly equal to the velocity vertical integral scale [i.e., $l_{F_{CO_2}}^{(z)} \approx l_w^{(z)}$].

7. First attempt to infer CO_2 turbulent flux by lidar

a. Efficient spectral filtering accounting for integral scale cutoff

Using the in situ data at 396 m, we studied an efficient filtering technique accounting for integral scales cutoff (from section 6) to limit the losses on useful information and for limited biases on CO_2 flux calculations.

The cross-spectrum of two different variables w and ρ_{CO_2} is defined as

$$\Phi_{F_{CO_2}}(\nu) = \frac{\Delta t}{M} \text{Re}[\text{DFT}_w(\nu/\Delta\nu, t) \text{DFT}_{\rho_{CO_2}}(\nu/\Delta\nu, t)^*], \quad (5)$$

where $\Phi_{F_{CO_2}}$ is the Fourier transform of the covariance between c and w (CCV) and $\text{DFT}_{\rho_{CO_2}}^*$ is the complex

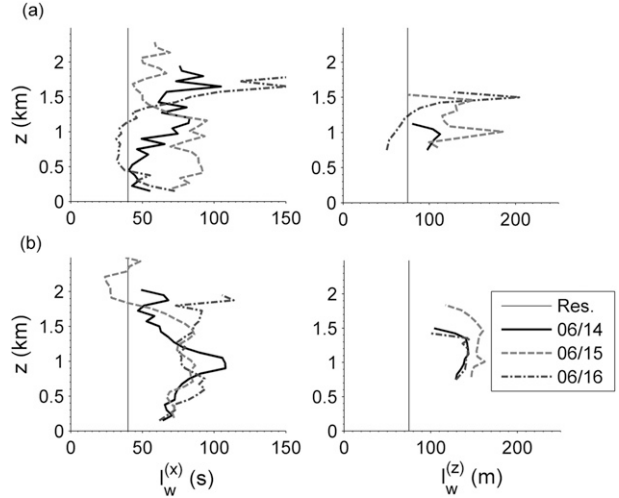


FIG. 6. Vertical profile of (left) temporal $l_w^{(t)}$ and (right) vertical $l_w^{(z)}$ velocity integral scales for (a) nights and (b) days using CDIAL data. Time (40 s) and space (75 m) CDIAL resolution are indicated with thin solid lines.

conjugate of $\text{DFT}_{\rho_{CO_2}}$. The integral of the cospectrum over the whole frequency range yields $\text{CCV}(0)$ (i.e., the vertical CO_2 flux).

Figure 7 shows the spectra and cospectra from D14 for CO_2 mixing ratio and vertical velocity time series using in situ data. The $-5/3$ inertial subrange is seen in CO_2 and velocity spectra. The frequency of the expected maximum contribution to the cospectrum is the inverse of the temporal CO_2 flux integral scale $l_{F_{CO_2}}^{(t)}$. Following Kristensen et al. (1989) we display the frequency ν_{\max} where the main contributions to covariance occur. It is linked to the flux integral scale as $\nu_{\max} = 1/[2\pi l_{F_{CO_2}}^{(t)}]$. There is little contribution to the total flux at wavelengths smaller than the integral scale. As shown in Giez et al. (1999) for water vapor flux, ν_{\max} is an upper bound for contribution of the cospectrum to CO_2 flux. Figure 7d shows that during daytime the time resolution of the lidar (i.e., 40 s) is smaller than the horizontal integral scale, and it is appropriate to get the main features of CO_2 flux in the whole CBL. However, the slow decrease of the integral of $\Phi_{F_{CO_2}}$ at high frequency entails significant biases on F_{CO_2} using low-time-resolution data, meaning that all the frequencies (i.e., eddies at different scales) contribute significantly to the daytime turbulent flux. For time averaging of 40, 80, and 160 s the corresponding bias is 16%, 29%, and 41%, respectively. These figures calculated at 396 m are used to correct the lidar CO_2 flux estimates. This will result in overestimation of the CO_2 flux because the biases are expected to be smaller in the middle of the CBL as $l_w^{(t)}$ is usually larger there than it is at 396 m (Fig. 6).

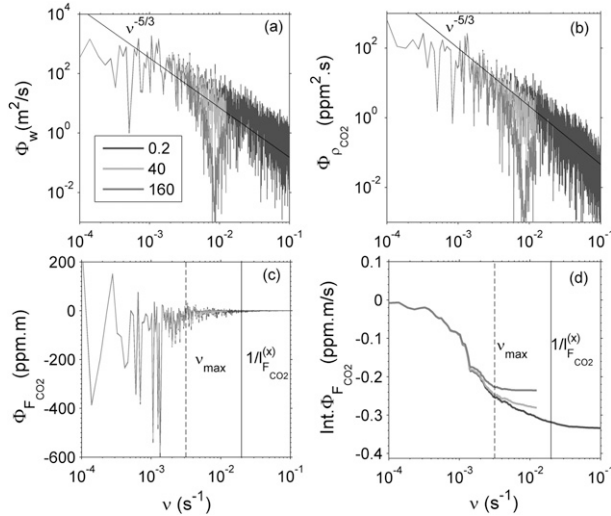


FIG. 7. The D14 power spectra of (a) in situ vertical velocity and (b) CO₂ mixing ratio at 396 m, between 1100 and 1300 LT. (c) Cospectrum and (d) integral of the cospectrum. Black, light gray, and gray lines are for 0.2-s data, 40-s averaged data, and 160-s rolling averaged data, respectively. The straight lines indicates the expected $\nu^{-5/3}$ law in the inertial subrange. The ν_{\max} marks the predicted maximum of the cospectral contribution to the flux. The inverse of the cross-covariance integral scale from Fig. 6a is also indicated.

b. First CO₂ flux measurements by lidar

The mean CO₂ fluxes are estimated in 2-h time gates for simultaneous CO₂ mixing ratio and vertical velocity time series at various range accumulation and time averaging [i.e., 150 m and 80 s (squares); 300 m and 160 s (stars)] in order to reduce as much as possible the statistical error of CO₂ mixing ratios. The results are displayed in Fig. 8. The CO₂ measurements are corrected for biases *C* and *D* [see Eq. (2)]. The 2-h flux estimates are then averaged over 6 h and 1200 m (for a range gate from 300 to 1500 m). Using in situ data we corrected the lidar CO₂ fluxes from biases due to H₂O flux [F_C in Eq. (3)] and lidar data averaging (section 7a).

During daytime, and despite large statistical errors, the lidar CO₂ flux estimates are significant and negative, as are the in situ fluxes. A CO₂ uptake by vegetation creates a sink in the surface layer that corresponds to a negative CO₂ flux at the bottom of the CBL. The free troposphere represents a source of CO₂ for the CBL as reported in section 4 and therefore we expect a negative flux of CO₂ at the top of the CBL.

During nighttime, the increase of CO₂ mixing ratio at the different levels below 244 m represents the buildup of CO₂ concentration in the nocturnal layer due to positive CO₂ surface flux associated with vegetation respiration (Fig. 8a). At the top of the residual layer, and because of larger free tropospheric CO₂ mixing ratio than in the residual layer, the entrainment flux is expected to be

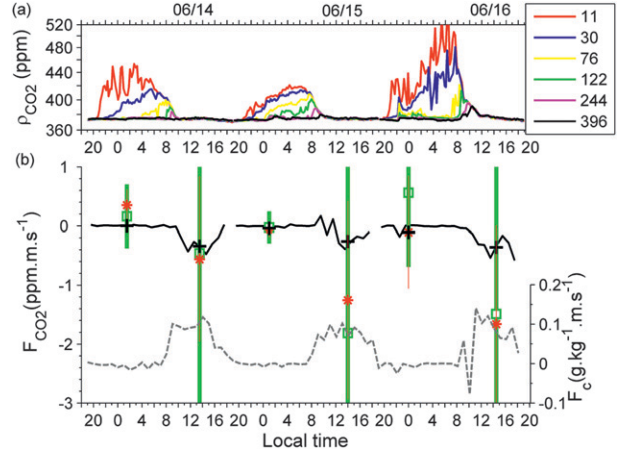


FIG. 8. (a) The CO₂ in situ mixing ratio measurements at 11, 30, 76, 122, 244, and 396 m. (b) In situ eddy covariance CO₂ flux at 396 m (black solid line) and 1.2-km ABL-mean lidar CO₂ flux estimates using the eddy covariance technique on 150-m, 80-s (green squares) and 300-m, 160-s (red stars) rolling averaged lidar CO₂ mixing ratio and vertical velocity measurements. The gray dashed line is for the in situ water vapor eddy covariance flux at 396 m used to correct H₂O bias on CO₂ flux measurements.

negative. These considerations explain the in situ negative CO₂ flux measured during N15 and N16. Lidar flux measurements using 300-m range and 160-s averaging are in good agreement with a global negative CO₂ flux in the whole RL (especially in the N15 and N16 cases when F_C is negligible).

From Eq. (3) we infer the random instrumental error in eddy flux measurements (Giez et al. 1999):

$$\sigma_{F_{CO_2, \text{inst}}}^2 = \frac{\Delta t \Delta z}{T Z} (\sigma_{w, \text{atm}}^2 \sigma_{\rho_{CO_2, \text{inst}}}^2 + \sigma_{\rho_{CO_2, \text{atm}}}^2 \sigma_{w, \text{inst}}^2 + \sigma_{\rho_{CO_2, \text{inst}}}^2 \sigma_{w, \text{inst}}^2), \quad (6)$$

where Δt and Δz correspond to the time and space resolution of lidar measurements and T and Z are respectively the time and space resolution of eddy covariance flux measurements.

The standard deviation of CDIAL flux estimates is approximately 850% during the daytime and 250% for N15 and N16 turbulent nights (Table 2). Large turbulent scales due to subsidence motion increase the lidar CO₂ flux estimates as a result of intermittent and large CO₂ fluctuations (between 0.5 and 2 ppm during N15 and N16). The experimental standard deviation calculated with the 2-h lidar CO₂ flux profiles and the theoretical standard deviation calculated with Eq. (6) give similar results. As a result, we are able to predict the performances of future instrument for accurate CO₂ flux measurements as discussed in section 8. Instrumental error due to online frequency jitter contributes to less than 1% of standard

TABLE 2. The 6-h averaged in situ and lidar eddy covariance flux measurements. Lidar flux estimates are averaged vertically over 1.2 km in the ABL. The lidar fluxes measurements are corrected from biases due to CDIAL data averaging (section 6a) and from water vapor flux (F_C). These biases are estimated using in situ data at 396 m. Standard deviations of CDIAL fluxes are calculated experimentally and theoretically using Eq. (7).

	N14	D14	N15	D15	N16	D16
In situ CO ₂ flux (ppm m s ⁻¹)	3.7×10^{-3}	-0.34	-0.04	-0.26	-0.11	-0.36
CDIAL CO ₂ flux (ppm m s ⁻¹)	0.35	-0.73	-0.06	-2.07	-0.09	-1.75
Space and time resolution of CDIAL data	150 m, 80 s	150 m, 80 s	300 m, 160 s	150 m, 80 s	300 m, 160 s	150 m, 80 s
Statistical error (ppm m s ⁻¹)						
Experimental	0.54	2.81	0.11	3.20	0.95	4.72
Theoretical	0.76	3.03	0.21	3.05	0.42	4.27
Corrected biases (ppm m s ⁻¹)						
F_C	0.190	-0.270	0.008	-0.255	0.020	-0.260
Signal averaging	-1.4×10^{-3}	+0.099	-5×10^{-4}	+0.098	-0.013	+0.109

deviation and is therefore negligible. A main source of random error in eddy covariance flux measurements is called the sampling error (Lenschow and Stankov 1986; Giez et al. 1999; Engelmann et al. 2008):

$$\sigma_{F_{\text{CO}_2, \text{samp}}}^2 = 2 \frac{l_{F_{\text{CO}_2}}^{(t)}}{T} (F_{\text{CO}_2}^2 + \sigma_{w, \text{atm}}^2 \sigma_{\rho_{\text{CO}_2, \text{atm}}}^2). \quad (7)$$

It depends on the number of dominant eddies or integral scales in the time interval T to calculate the CO₂ flux. In our cases, T is quite large (>2 h) so this error is negligible and below 0.5%.

The bias F_C due to water vapor absorption is corrected. Notice that it contributes to nearly half the lidar CO₂ flux estimates. This outlines the importance of proper location for the on- and off-laser lines. The corrected biases due to signal averaging (section 7a) range between 30% and 40% during daytime and are lower than 10% for N15 and N16.

8. Evaluation of future 2- μm CDIAL for accurate CO₂ turbulent fluxes

a. Requirements on CO₂ and velocity measurements

Signal averaging can be used to decrease the instrumental statistical error on CO₂ flux measurements as long as the conditions on time and space integral scales of CO₂ turbulent flux [i.e., $\Delta z < l_{F_{\text{CO}_2}}^{(z)}$ and $\Delta t < l_{F_{\text{CO}_2}}^{(t)}$] are fulfilled to result in negligible biases with regard to flux measurements. The lower bound for lidar vertical range resolution is 150 m (Fig. 6). To estimate a lower bound for lidar time accumulation we use in situ data at 396 m to calculate CO₂ turbulent flux with different time resolution in a T -time gate of 2 h (Fig. 9). The 2-h time gate is chosen as a compromise to have a negligible sampling error [see Eq. (7)] and to keep relevant flux measurements during the morning or evening transition. Similar results have been obtained during daytime. A smaller time averaging than 10 s enables us to keep the biases

below 10%. It is taken as a requirement for lidar CO₂ flux measurements. A lower time resolution can be used during windy nights, especially during the N15 case when the bias is lower than 4% for a 1-min averaging.

To obtain a lower bound for lidar CO₂ and velocity standard deviation, we add CO₂ and velocity random error variances on 10-s averaged in situ data and calculate the resulting standard deviation for 2-h average flux estimates. Figure 10 displays the results for the N14 and D14 cases. During daytime, Fig. 10b shows that limited CO₂ mixing ratio precision as low as 15 ppm (i.e., 50 times the CO₂ turbulent fluctuations reported in section 5 for in situ probes, with $\Delta t < 10$ s and $\Delta z < 150$ m) is sufficient to reach an F_{CO_2} relative standard deviation lower than 50%. Here, 50% is taken here as an arbitrary bound for relevant CO₂ flux measurement. During the night, this precision will also be sufficient for highly turbulent conditions (N15 and N16) similar to those observed during daytime. For N14 stable conditions in the residual layer, Fig. 10a shows that a 2-ppm precision is required (i.e., 6 times the CO₂ fluctuations reported by in situ probes).

The instrumental error on vertical velocity has a negligible impact on eddy covariance flux error as long as the noise standard deviation remains lower than the natural standard deviation (i.e., 0.1 and 1 m s⁻¹ for N14 and D14, respectively).

b. Dimension of a future 2- μm CDIAL for accurate CO₂ fluxes

The precision of the CO₂ mixing ratio relies on the number of independent samples in a range gate (notice the coherence time is limited by the laser pulse duration; see Bruneau et al. 2006), whereas the accuracy with regard to velocity relies on the precision in the spectrum of the heterodyne signal. For simultaneous velocity and concentration measurements, a compromise is needed for similar accuracies in the time and spectral domains. Figure 10 shows that accuracy in CO₂ mixing ratio measurements

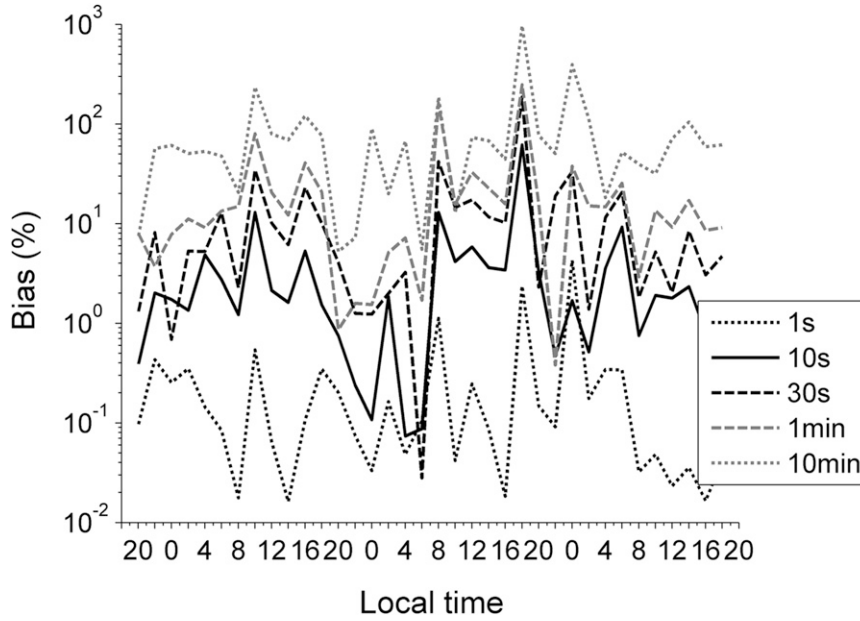


FIG. 9. Bias on 2-h mean CO₂ flux estimates for different time resolutions of in situ data at 396 m.

demands short duration pulses (<100 ns) to increase the number of samples and accuracy in the time domain.

Using the lidar equation and assuming the same off- and on-line instrumental constant K , energies E , heterodyne efficiencies γ_H , backscatter coefficient β , and atmospheric transmission T_{at} except for the CO₂ absorption (T_{CO_2}), we can write (Gibert et al. 2008)

$$CNR_{on} = CNR_{off} T_{CO_2}^2, \quad (8)$$

$$CNR_{off} = KE\gamma_H\beta T_{at}^2/z^2. \quad (9)$$

The CO₂ lidar standard deviation at $z + \Delta z/2$ can be calculated by

$$\sigma(\rho_{CO_2})/\rho_{CO_2} = [SNR_{on}^{-2}(z) + SNR_{off}^{-2}(z) + SNR_{on}^{-2}(z + \Delta z) + SNR_{off}^{-2}(z + \Delta z)/(2n_{air}\Delta\tilde{\sigma}_{CO_2}\Delta z)]^{1/2}, \quad (10)$$

where the SNR (which accounts for speckle and detection noise) is calculated from experimental CNR, using an analytical expression from Rye and Hardesty (1997):

$$SNR = \sqrt{M_s M_t} / (1 + CNR^{-1}). \quad (11)$$

Here $M_t = \sqrt{1 + (\Delta t/T_c)^2}$ is the number of coherence cells in a range gate, assuming a Gaussian pulse of duration T_c and a rectangular range gate of duration Δt , and M_s is the number of shot-pair on-offs.

Numerical simulations were performed to estimate the precision of the CO₂ mixing ratio as a function of the number of independent samples $M_t M_s$. The lidar time and range resolution are 10 s and 150 m, respectively. Figure 11 displays the results at $z = 1$ km. Here T_{at} and T_{CO_2} are calculated using standard meteorological atmosphere profiles and new spectroscopic data for the R30 CO₂ absorption line (Joly et al. 2009). The R30 line

is more favorable than the R22 line used in the experiments described here. The on-laser line is located at the center of the CO₂ absorption line. The laser pulse duration is 50 ns, which entails $M_t \sim 20$. The heterodyne efficiency is 10% and $\beta = 5 \times 10^{-7} \text{ m}^{-1} \text{ sr}^{-1}$ so that we can display additional E -PRF axes corresponding to the CNR - $M_t M_s$ ones in Fig. 11. Assuming a sequential emission of on- and off-lines, the bold $\sigma(\rho_{CO_2}) = 10$ ppm line is reached for a laser PRF > 2 kHz or $E > 2$ mJ, and a mean power PRF $\times E > 24$ W. In addition, Fig. 11 shows that $E < 10$ mJ and ERF > 4 kHz are more favorable to decrease the total power of the laser emitter.

9. Conclusions

In this paper the requirements for accurate range-resolved CO₂ flux measurements in the atmospheric boundary layer by combined and simultaneous DIAL

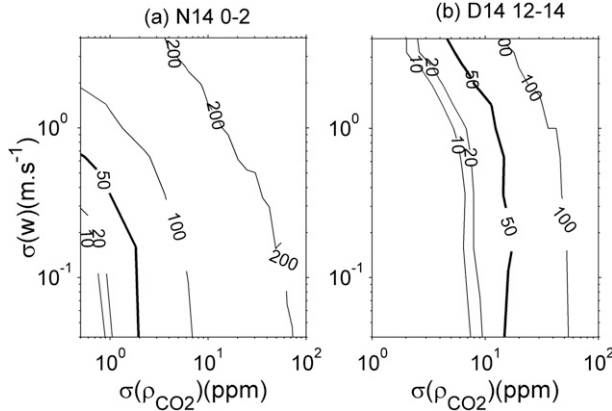


FIG. 10. The CO_2 turbulent flux error (percent) as a function of vertical velocity and CO_2 mixing ratio measurement standard deviations for 10-s time resolution and for (a) N14 and (b) D14. Flux time resolution is 2 h.

and Doppler lidar techniques are quantitatively defined in terms of accuracy and time and space resolution. It is shown that the turbulent fluctuations of CO_2 mixing ratio amount to ~ 0.3 and 0.5 ppm during night and day, respectively, whereas the vertical velocity fluctuations are around $\sim 0.1 \text{ m s}^{-1}$ and 1 m s^{-1} during night and day, respectively. Time and space resolution of the lidar measurements is constrained (i.e., limited) by relevant time and space integral scales that are determined by the size of coherent structures of turbulence. Despite a horizontal integral scale larger than 50 s in the convective BL, it is found that shorter lidar time resolution (i.e., < 10 s) is needed to avoid significant biases on CO_2 flux estimates. During the night, the integral scale varies over two orders of magnitude, from 3 s to few minutes, depending on the occurrence of nocturnal jet or subsidence motion. The vertical integral scale is around 150 m in the CBL and can be lower than 100 m during the night. It is shown that the horizontal and vertical integral scales are the longest in the middle of the CBL.

Despite the limitation with regard to CO_2 mixing ratio measurements by lidar compared to turbulent fluctuations (as reported by in situ sensors), we were able to report preliminary mean lidar CO_2 flux measurements in the ABL with time and space resolutions of 6 h and 1200 m, respectively. Depending on the turbulent fluctuations of CO_2 mixing ratio, the CO_2 flux error by CDIAL ranges between 250% for jet-disturbed nights and 850% during the day. These large uncertainties are reported for current CDIAL, which is not designed for the purpose of CO_2 flux retrieval. Biases due to parasitic water vapor absorption and signal averaging were corrected using in situ data. This drawback can be avoided for CDIAL instruments properly designed for CO_2 flux measurements. In this respect, numerical simulations

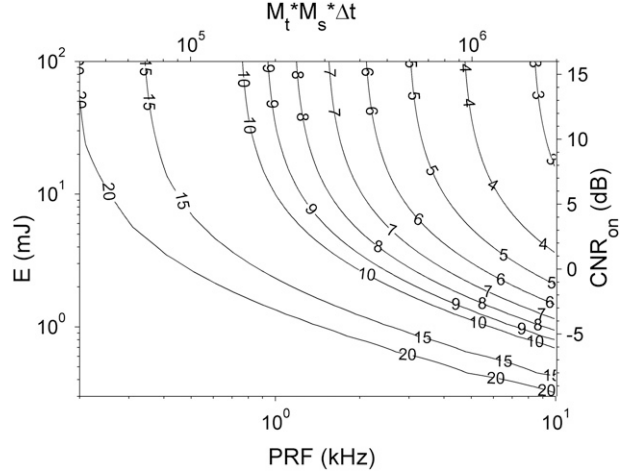


FIG. 11. The CDIAL 10-s, 150-m averaged CO_2 mixing ratio standard deviations (ppm) as a function of on-line CNR (dB) or laser pulse energy (mJ) and the number of independent samples or laser pulse repetition frequency (kHz).

using in situ CO_2 and velocity measurements are conducted for the design of future CDIAL instrument. Although the turbulent fluctuations of CO_2 mixing ratio are only a small portion of the mean value (0.1%), we show that a 25-W $2\text{-}\mu\text{m}$ CDIAL operating with a 2-h time integration and 150-m range resolution can retrieve accurate CO_2 flux estimates (better than 50%) as long as the CO_2 mixing ratio instrumental error is smaller than 10 ppm and the vertical velocity error is lower than fluctuations over a time period of 10 s.

Acknowledgments. This research was funded by the NASA Instrument Incubator Program and NASA Laser Risk Reduction Program. We thank R. Strand and J. Ayers of the Wisconsin Educational Communications Board for hosting the lidar at the WLEF tower.

APPENDIX A

Similarity Laws to Predict Vertical Velocity and Scalar Variances in the CBL

The similarity law for vertical velocity variance in the CBL can be calculated using Stull (1988):

$$\overline{w'^2} = 1.8w_*^2(z/z_i)^{2/3}(1 - 0.8z/z_i)^2, \quad (\text{A1})$$

where $w_* = (gz_i\overline{w'\theta'_0}/\theta)^{1/3}$ is the free-convection scaling velocity, $\overline{w'\theta'_0}$ the in situ surface heat flux (calculated at 30 m), θ the mean potential temperature in the CBL, and z_i the top of the CBL.

The theoretical turbulence variance of a scalar concentration like CO_2 in the CBL can also be estimated using Moeng and Wyngaard (1984):

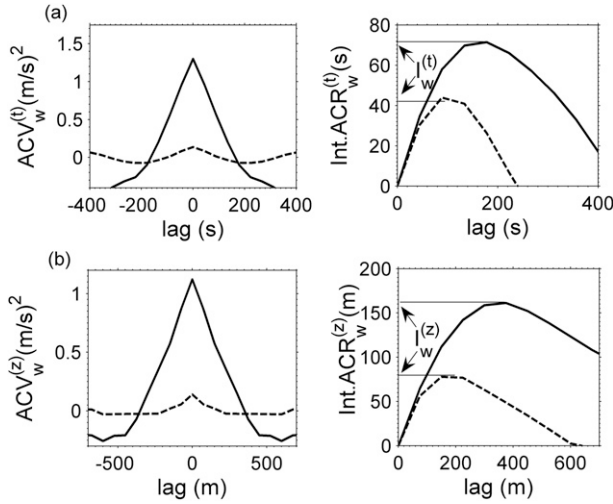


FIG. B1. (a) (left) D14 (solid line) and N14 (dashed line) time autocovariance function of CDIAL vertical velocity at 375 m ($ACV_w^{(t)}$) and (right) integrals of the autocorrelation functions ($Int.ACR_w^{(t)}$). The time-integral scale is defined as the maximum of the integral (b) As in (a), but for the vertical integral scale calculation.

$$\overline{\rho_{CO_2}^2} = (\overline{w' \rho_{CO_2}'} / w_*)^2 f(z/z_i), \quad (A2)$$

where ρ_{CO_2}' is the in situ CO_2 surface flux (here calculated from the 30-m level of the WLEF tower) and $f(z/z_i)$ is the bottom-up variance function estimated from large-eddy simulation (LES) and equal to $(z/z_i)^{-0.9}$ for $z > 0.1z_i$ and $1.8(z/z_i)^{-2/3}$ for $z < 0.1z_i$.

APPENDIX B

Turbulence Time and Space Scales—Methodology

In this section, we investigate the turbulence characteristics of in situ and lidar observations using covariance techniques. The autocovariance (ACV) is used to separate signal variance due to space-correlated atmospheric processes from uncorrelated instrumental noise. For the atmospheric variable $c(x)$,

$$ACV_c(X) = \overline{c(x)'c(x+X)'}, \quad (B1)$$

where $c(x) = \bar{c} + c'(x)$, with $c'(x)$ being the space-dependant fluctuation and X the space lag. The overbar represents the mean in the range gate used to calculate the autocovariance.

For lidar measurements, Eq. (B1) becomes $ACV_c(X) = \langle c(x) \rangle' \langle c(x+X) \rangle'$, where the angle brackets indicate both time and space lidar averaging. Since we used a ground-based instrument, the horizontal ACV is a function of time. Knowing that $ACV_c(0) = \sigma_c^2 =$

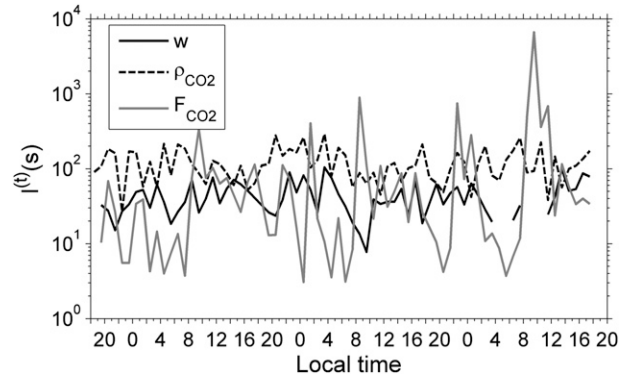


FIG. C1. Hourly time integral scale of vertical velocity, CO_2 mixing ratio, and CO_2 flux at 396 m.

$\sigma_{c,inst}^2 + \sigma_{c,atm}^2$, and using a Fourier transform to determine $\sigma_{c,inst}^2$ (Fig. 3), we can measure $\sigma_{c,atm}^2$. Then we can define the autocorrelation function (ACR) as $ACR_c(X) = ACV_c(X) / \sigma_{c,atm}^2$. The integral of this function is called the integral scale (IS) (Lenschow and Stankov 1986):

$$l_c = \int_0^\infty ACR_c(r) dr. \quad (B2)$$

The first maximum of this integral is usually chosen to be the IS (Lenschow and Stankov 1986; Giez et al. 1999; Lothon et al. 2006). The IS is related to the dominant eddy size and enables us to determine the space and time scales of turbulence. A flux integral scale can also be defined using the covariance of two measured variables c and w :

$$CCV(X) = \overline{c(x)'w(x+X)'}, \quad (B3)$$

where $CCV(0)$ corresponds to the vertical flux of $c(x)$.

Figure B1 shows the time and space velocity autocovariance function and the corresponding integrals of autocorrelation functions. We use the maximum of the integral of $ACR_w^{(t)}$ to define $l_w^{(t)}$. We use the same method to calculate $l_w^{(z)}$ rather than the exponential fit method used in Lothon et al. (2006), which was found to be less accurate in our cases.

APPENDIX C

Comparison of Turbulence Integral Scales of Vertical Velocity, CO_2 Mixing Ratio, and CO_2 Flux

Figure C1 shows a comparison of time integral scale for vertical velocity, CO_2 mixing ratio, and CO_2 flux using the method detailed in appendix B. During the day, $l_{\rho_{CO_2}}^{(t)}$, $l_w^{(t)}$, and $l_{F_{CO_2}}^{(t)}$ take similar values, which shows

that eddies of the predominant size direct both velocity and scalar fluctuations in the CBL. During the night, no previous observations in the residual layer can help us to understand the large integral scales that we see for the CO₂ mixing ratio. For N15 case, the nocturnal jet seems to explain similar increases in $l_w^{(t)}$ and $l_{\rho_{\text{CO}_2}}^{(t)}$. An important result from Figure C1 is that $l_{F_{\text{CO}_2}}^{(t)}$ varies from 5 to 1000 s with large variations from one hour to another, especially during the night and early morning. Large temporal scale correlation exists between CO₂ mixing ratio and vertical velocity during the morning transition when z_i reaches each level of the tower and when the overshoots bring the large CO₂ mixing ratio from the former NBL into the RL. This is in good agreement with previous observations that conclude that entrainment plays a major role in CBL CO₂ drawdown during the morning transition (Vila-Guerau de Arellano et al. (2004); Gibert et al. 2007b).

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